NATIONAL OCEANIC AND ATMOSPHERIC ADMINISTRATION NATIONAL WEATHER SERVICE SYSTEMS DEVELOPMENT OFFICE TECHNIQUES DEVELOPMENT LABORATORY

TDL OFFICE NOTE 82-2

RECENT EXPERIMENTS IN THE USE OF MODEL OUTPUT STATISTICS FOR FORECASTING SNOW AMOUNTS

Joseph R. Bocchieri

1. INTRODUCTION

precipitation (PoF) (Bocchieri, 1979b). National Weather Service, 1980), and the conditional probability of frozen probability, the probability of precipitation (PoP) (Lowry and Glahn, 1976; unconditional probability is derived by taking the product of the conditional ability of heavy snow and a categorical forecast are also provided. The periods after 0000 GMT and 1200 GMT. An estimate of the unconditional proba statistical relationship is determined between the forecast output of a operational within the National Weather Service since October 1977 (Bocchieri tion equations that give the conditional probability of heavy snow for 12-24 h from both the Limited-area Fine Mesh (LFM) model (National Weather Service, numerical prediction model 24 inches during a 12-h period at a station. In MOS (Glahn and Lowry, 1972), 1979a); National Weather Service, 1978). Heavy snow is defined as a fall of (National Weather Service, 1977; Newell and Deaven, 1981), to develop prediclar weather element. For the operational heavy snow system, we used output 1971; Gerrity, 1977) and a finer mesh version of the LFM, called LFM-II A Model Output Statistics (MOS) system for forecasting heavy snow has been (or models) and observed occurrences of a particu-

tions and found that 12 predictors was about right. determined the optimum number of predictors to include in the forecast equations that included boundary-layer predictors. include boundary-layer predictors were only slightly less accurate than equa-Meteorological Center may not include a boundary layer in the form presently cast equations, since future operational numerical models at the National mined the effect of removing LFM boundary-layer type predictors from the foreequations, we performed a number of experiments. In one experiment, we detercasts for other snow amount categories, not just heavy snow, and other forewith the goal of developing a new operational system which would provide forecast projections other than 12-24 hours. In the process of deriving these We've recently derived a set of experimental snow amount forecast equations The results indicate that forecast equations that didn't In another experiment, we

didn't include boundary-layer type predictors. of the two systems. system; the results indicate that there was little difference in the accuracy equations were then compared to similar forecasts made by the Conditional probability of heavy snow forecasts made by the experimental This was encouraging because the experimental equations operational

amount categories, given the conditional probabilities from the experimental methods for estimating the unconditional probability of occurrence of the snow conditional probabilities for each snow amount category by the Pop for the 12-18, 18-24, and 24-36 h projections from 0000 GMT. We then compared two conditional probability equations for several snow amount categories for the After performing these experiments for the 12-24 h projection, we developed One method, called PRODUCT, consisted simply of multiplying the

procedure can be found in Glahn and Lowry (1972). between the predictors and predictands. A good description of the screening predictors helps to account for non-linear relationships which may exist Table I were included in both binary and continuous form. The use of binary

a. Preliminary Experiments

The two PoSA(S) systems forecast the probability of ≥ 4 inches of snow, given that snow occurs, in the 12-24 h period after 0000 GWT. potential predictors in Table 1 to develop 12-term equations for each system. other system didn't. 1978-79. One system included boundary-layer predictors, among others, and the stations and seven winter seasons, September through April, 1972-73 through systems were developed with data combined from 195 conterminous United States boundary-layer predictors from the PoSA(S) equations. Two PoSA(S) forecast number of experiments. We first determined the effect of removing LFM the process of developing the experimental PoSA(S) equations, we did The REEP technique was used in conjunction with

tions, since these fields may not be available from future numerical models to include boundary-layer predictors in further development of PoSA(S) equapendent data from the winter season of 1979-80. The results show that the run at the National Meteorological Center. predictors was about 1% better than the Brier score of the system that didn't Brier score (Brier, 1950) for the PoSA(S) system which included boundary-layer have these predictors. We judged that this improvement was not large enough We did a comparative verification between the two PoSA(S) systems with inde-

ing manner. We derived PoSA(S) equations for the ≥ 1 and ≥ 4 inch categories for the 12-24 h period from both 0000 GMT and 1200 GMT with data combined from was computed for each station and for each snow amount category by: on the developmental sample. A statistic called the relative probability bias approach. 195 stations for the developmental sample -- the so-called generalized operator seasons (1972-73 through 1979-80). The regions were determined in the followdevelopmental sample consisted of data from 195 stations and eight winter 12-24 h period after 0000 GMT for each of several geographic regions. The include in PoSA(S) equations. First, PoSA(S) equations were derived for the In another experiment, we determined the optimum number of predictors to We then evaluated the equations to obtain forecasts for each station

relative probability bias characteristics; other factors we considered were determined the regions shown in Fig. 1 by grouping stations having similar category for each station for the developmental sample. We subjectively snow amount category for each station and RF is the relative frequency of that cases would be available for equation development. to make the regions large to insure that a sufficient number of snow amount the density of stations and the climatic frequency of snow amount. We needed where PoSA(S) is the average conditional probability forecast for a particular

various parameters at 500-mb to be useful. 10-5 THICK ADV ranked relatively also found the 850-mb level to be useful for snow amount prodiction. and Younkin (1970), Brandes and Spar (1971), and Spiegler and Fisher (1971) ranked next in importance at 24-36 hours. Other investigators such as Browne 850 M CONV, ranked next in importance at 12-24 hours, and 850 U and 850 DIV Various parameters derived at 850-mb, such as 850 DIV, 850 VORT, and of this ranking, all predictor prjections, smoothings, and binary limits were list was determined by both frequency and order of selection; for the purpose ing procedure for the 12-24 h and 24-36 h projections from 0000 GMT. This high for the 12-24 h projection; Younkin (1968) found strong warm advection to followed by 700 U. Goree and Younkin (1966) and Weber (1978) also found for 12-24 hours. Variables at 700-mb, such as 700 U and 700 VV followed the 850-mb parameters P AMT and MEAN R HUM were the most important predictors for both projections. combined for each type of variable. The results indicate the LFM forecasts of be associated with heavy snow in the western United States. In Table 4, the predictor types are ranked as determined by the REEP screen-For the 24-36 h projection, the 500 VORT ranked next

3. EXPERIMENTS IN ESTIMATING THE UNCONDITIONAL PROBABILITY OF SNOW AMOUNT

corresponding 12-h period and the average PoF (\overline{PoF}) for the same 12-h period. This is expressed mathematically¹, for instance for the ≥ 1 inch category, by: ability forecast, PoSA(S), for each snow amount category, by the PoP for the method, called PRODUCT, consists simply of multiplying the conditional prob-We experimented with two methods for obtaining unconditional probability of snow amount, PoSA, forecasts for the 12-24 h period after 0000 GMT. One

$$PoSA(\geq 1 \text{ inch}) = PoSA(S)(\geq 1 \text{ inch}) \times PoP \times PoF.$$
 (2)

estimates of the unconditional probability of heavy snow. This method is essentially the one used in the operational system to obtain

PoSA(S) equations in Section 2. That is, we computed the relative probability combined from 192 stations for the ≥ 1 , ≥ 2 , ≥ 3 , ≥ 4 , and ≥ 6 inch categories. It then attempted to determine regions in a manner similar to that used for the PoF forecasts, and various products thereof. The developmental sample for predict PoSA. Potential predictors for these equations were PoSA(S), PoP, and this derivation consisted of the months October through March, 1972-73 through to combine stations with similar bias characteristics to determine the regions bias for each station by Eq. (1) using the developmental sample and then tried The other method, called MOSSQR, involved derivation of REEP equations to We first developed generalized-operator PoSA equations with data

In a true mathematical sense, this method doesn't give the unconditional probability of snow amount for 12-h periods in which mixed precipitation occurs, since PoSA(S) equations were developed with pure snow cases only.

forecast for that category would result in a categorical forecast of ≥ 1 inch-If the threshold value is not exceeded, the categorical forecast would be

provided. For subsequent iteration, the threshold probability was incremented forecasts made by comparing probability forecasts against threshold probabilibias was <1.30, even if the maximum threat score was associated with a bias Threshold probabilities were computed with the use of an empirical iterative preselected value. For the initial iteration, a first guess threshold probability was On each iteration, threat scores were computed for categorical Threshold values were chosen so that the categorical

gory. Since it doesn't, the procedure is terminated and the forecast amount 0.23, 0.16, and 0.08. The procedure starts at the category ≥ 1 inch and compares the probability forecast (0.40) with the threshold value for that is 1 to 2 inches. predicted. The next step is to proceed to the category >2 inch and see if the probability forecast (0.30) exceeds the threshold value (0.32) for that cateand 0.05, respectively. for the categories ≥ 1 , ≥ 2 , ≥ 3 , ≥ 4 , and ≥ 6 inches are 0.40, 0.30, 0.20, 0.15, and 0.05, respectively. Furthermore, assume that the threshold probabilities value for the category >1 inch, the forecast amount would have been <1 inch. is 1 to 2 inches. Of course, if the probability forecasts had exceeded the threshold values for all categories, the forecast amount would have been category (0.38). Since the threshold value is exceeded, at least 1 inch is that maximize the threat score for each of the five categories are 0.38, 0.32, As as example, assume that the probability forecasts for the 12-24 h period Similarly, if the probability forecast was less than the threshold

gories of snow amount, $\frac{>1}{>1}$, $\frac{>2}{>2}$, $\frac{>3}{>2}$, $\frac{>4}{>2}$, and $\frac{>6}{>3}$ inches; for the two 6-h periods, we verified three categories, $\frac{>1}{>1}$, $\frac{>2}{>2}$, and $\frac{>3}{>3}$ inches. The results indicate the projections from 0000 GMT. data samples, which were the same as those used in Section 3. categorical snow amount forecasts for both the developmental and independent following: the threat score and bias for the 12-24, 12-18, 18-24, and 24-36 h forecast After determining the threshold probability values, we produced and verified For the two 12-h periods, we verified five cate-Table 7 shows

- scores were between .20 and .30 for the ≥ 1 , >2, and ≥ 3 inch categories for the 12-24 and 24-36 h projections and for the ≥ 1 and For the developmental sample, the biases were generally between generally between .15 and .20 for the remaining categories for each >2 inch categories for the two 6-h periods. The threat scores were probability values were derived. For these bias values, the threat 1.00 and 1.30. The bias was forced to be such when the threshold
- was worse for the 18-24 h and 24-36 h periods than for the 12-18 h and 12-24 h periods. Also, the deterioration was generally worse For the independent sample, both the threat score and bias deteriorated as compared to the developmental sample. The deterioration

experimental PoSA(S) equations for other projections. operational equations did. We therefore proceeded with the development of mental equations didn't include boundary-layer type predictors while the accurate as the operational forecasts, in spite of the fact that the experi-

moisture convergence, and wind components ranked next in importance. Similar parameters derived at the 850-mb level, such as divergence, vorticity, projections from 0000 GMT. An analysis of the predictor types chosen by the variables at the 700-mb and 500-mb levels followed. humidity were the most important predictors for both projections. LFM forecasts of precipitation amount and surface to 500-mb mean relative REEP screening procedure for the 12-24 h and 24-36 h projections showed that PoSA(S) equations were developed for the 12-18 h, 18-24 h, and 24-36 h

of regression equations to p edict PoSA. Potential predictors for these equations were PoSA(S), PoP, and PoF forecasts, and various products thereof. A each snow amount category by the PoP for the corresponding 12-h period and the average conditional probability of frozen precipitation, PoF, for the same method, called PRODUCT, consists simply of multiplying PoSA(S) forecasts for of snow amount, PoSA, forecasts for the 12-24 h period after 0000 GMT. One we decided to use the PRODUCT method to estimate PoSA, since it will be easier cates there was little difference in Brier scores for the two systems. Hence, comparative verification between PRODUCT and MOSSQR on independent data inditional heavy snow system. Another method, called MOSSQR, involved derivation 12-h period. This is quite similar to the method used in the present operato put into operation. We also expermented with two methods for obtaining unconditional probability

snow amount forecasts for the independent data sample. The results indicate developmental and independent data samples. The PoSA forecasts were then method to estimate PoSA for each projection, each region, and for both the gorical snow amount forecasts on independent data. First, we used the PRODUCT would be suitable for possible operational implementation, we verified catecompared to the lower categories for each forecast period. We also compared ration in the scores was generally worse for the higher snow amount categories 24-36 h periods than for the 12-18 h and 12-24 h periods. Also, the deteriofor the developmental sample. The deterioration was worse for the 18-24 h and that both the threat score and bias deteriorated as compared to these scores threshold probability values for each snow amount category, each region, and transformed into categorical snow amount forecasts; to do this, we derived results indicated that the two systems were of comparable accuracy. to the same scores computed for the operational heavy snow system. the threat score and bias for the >4 inch category for the experimental system To help us determine whether the experimental snow amount forecast system to <1.30. We then computed the threat score and bias for categorical chosen such that the threat score was maximized while restricting the forecast projection using the developmental sample. The threshold values

- HOWLY, D. A., and H. H. atamin, probability of precipitation--PEATMOS PoP. Mon. Wea. Rev., 104, 221-231.
- Miller, R. G., 1964: Regression estimation of event probabilities. Inc. Tech. Rep. No. 1, Contract CWB-10704, The Travelers Research Center, Hartford, Conn., 153 pp. [NTIS AD 602037].
- National Weather Service, 1971: Tech. Proc. Bull., No. 67, NOAA, U.S. Department of Commerce, 11 pp. The limited-area fine mesh (LFM) model. NWS
- NOAA, U.S. Department of Commerce, 6 pp. 1977: High resolusion LFM (LFM-II). NWS Tech. Proc. Bull., No. 206.
- of heavy snow. NWS Tech. Proc. Bull., No. 246, NOAA, U.S. Department of Commerce, 10 pp.
- precipitation. Commerce, 13 pp. 1980: The use of model output statistics for predicting probability of tation. NWS Tech. Proc. Bull. No. 289, NOAA, U.S. Department of
- Newell and Deaven, 1981: NOAA, U.S. Department of Commerce, 20 pp. The LFM-II model--1980. NOAA Tech. Memo. NWS NMC-66,
- Spiegler, D. B., and G. F. Fisher, 1971: A snowfall prediction method for the atlantic seaboard. Mon. Wea. Rev., 99, 311-325.
- Weber, E. M., 1978: Major midwestern snowstorms. Preprints Conference on Meteorological Society, Silver Spring, 30-37. Weather Forecasting and Analysis and Aviation Meteorology, American
- Younkin, R. J., 1968: Circulation patterns associated with heavy snowfall over the western United States. Mon. Wea. Rev., 96, 851-853.
- Zurndorfer, E. A., and R. J. Bermowitz, 1976: Determination of an optimum number of predictors for probability of precipitation amount forecasting. TDL Office Note 76-17, National Weather Service, NOAA, U.S. Department of Commerce, 7 pp.

20	16	100	10	σ	σ	4	N	·	Predictors	Number of	
.384	.383	. 385	•386	.387	.388	. 389	.395	.406	١×		
.229	. 229	.230	.229	.230	. 233	.232	.232	.232	>2	Sno	
.131	•131	.132	.131	.132	.132	.133	.133	.133	<u>></u> 3	ow Amount Ca	Brier
.071	.072	.071	.071	.071	.071	.073	.073	.072	I>4	Snow Amount Category (inches	Brier Scores
.023	.023	.023	.023	.023	.022	.022	.023	.023	>6	es)	
							100				

Table 3. The reduction of variance for REEP conditional probability of snow amount equations for the 12-24 h projection from 0000 GMT for each of the five regions shown in Fig. 1. The developmental data were from 8 winter seasons (1972-73 through 1979-80). The relative frequency (%) of each snow amount category is also shown in parentheses.

ъ	4	U	N	8 -3 1	Region	
22.8 (36.4)	18.0 (39.4)	17.8 (36.7)	19.2 (49.3)	10.9 (37.3)	17.	
24.1 (17.8)	24.9 (19.6)	17.9 (17.9)	15.8 (27.9)	9.9 (16.9)	Snow Amou	Total Re
22.5 (9.8)	20.6 (11.1)	17.9 (9.2)	14.5 (17.8)	7.3 (7.4)	Snow Amount Category (inches) >2 >3 >4	Total Reduction of Variance (%)
20.8 (5.9)	18.9 (6.4)	18.3 (4.6)	11.6 (10.3)	8.6 (3.1)	y (inches)	Variance
17.3 (2.5)	18. ² (2.3)	13.2 (1.4)	6.8 (3.4)	7.4 (0.9)	>6	(%)
3799	3785	4170	1100	1307	Total Number of Snow Cases	14 148 (1997)

reduction of variance given by each predictor is shown. The number of cases for each snow amount category is given in parentheses.

	Addit	Additional Redu	Reduction of V	Variance (%)	3
Predictors	ا <u>ن</u> دا	Snow Amount	Category >3	(inches) <u>>4</u>	1>6
1 3	(6599)	(3170)	(1679)	(899)	(307)
PoSA(S) (>1 in) x PoP x PoF	27.4	19.8	13.7	9.3	4.5
PoSA(S) (>6 in) x PoP x PoF	0.1	1.5	2.8	4.5	7.2
PoSA(S) (>2 in) x PoP x Pof <70%	0.0	0.0	0.1	0.6	1.4
PoSA(S) (>3 in) x PoP x PoF	0.0	0.5	0.8	0.5	0.0
$PoSA(S)$ (≥ 4 in) x PoP x PoF $\leq 50\%$	0.0	0.0	0.0	0.0	0.5
$PoSA(S)$ (≥ 2 in) x PoP x PoF $\leq 50\%$	0.0	0.0	0.1	0.3	0.1
$PoSA(S)$ (≥ 6 in) x PoP x PoF $\leq 4.5\%$	0.0	0.0	0.0	0.0	0.2
PoSA(S) (≥2 in) x PoP x PoF ≤18%	. 0.1	0.2	0.1	0.1	0.1
$PoSA(S)$ (≥ 4 in) x PoP x PoF	0.0	0.0	0.0	0.2	0.0
Climatic Frequency >6 in	0.0	0.0	0.0	0.1	0.1
$PoSA(S)$ (≥ 6 in) x PoP x PoF $\leq 1.5\%$	0.0	0.0	0.0	0.0	0.1
PoSA(S) (≥2 in) x PoP x PoF <80%	0.0	0.0	0.0	0.0	0.1

	24-36 h	18-24 h	12-18 h	12-24 h	Forecast Projection from 0000 GMT
9	Threat score Bias Number of snow cases	Threat score Bias Number of snow cases	Threat score Bias Number of snow cases	Threat score Bias Number of snow cases	Verification Score
	.28 1.29 4387	.26 1.22 2980	.27 1.10 3470	.30 1.08 6599	D >1
	.22 1.09 655	.19 .63 305	. 26 . 77 306	.28 .83 629	I
	.22 1.14 2082	.21 1.05 1212	.21 1.13 1353	.25 1.10 3174	Snow
	.13 .75 278	.11 .39 119	.16 .71 126	.19 .65 298	Amou
	.20 1.19 1108	.18 1.20 528	.17 1.02 584	.21 1.07 1671	Snow Amount Category (inches) 2 I D 3 I D 4
	.57 133	.12 .40 47	.10 .56	.14 .61 150	gory (
	.17 1.16 564	111	111	.19 1.09 902	inches)
	.00 .25	[]]	111	.64 83	н
	1.11 1.11 185	111	111	.18 1.03 310	D >6
	.00 .14 21			1.00	н

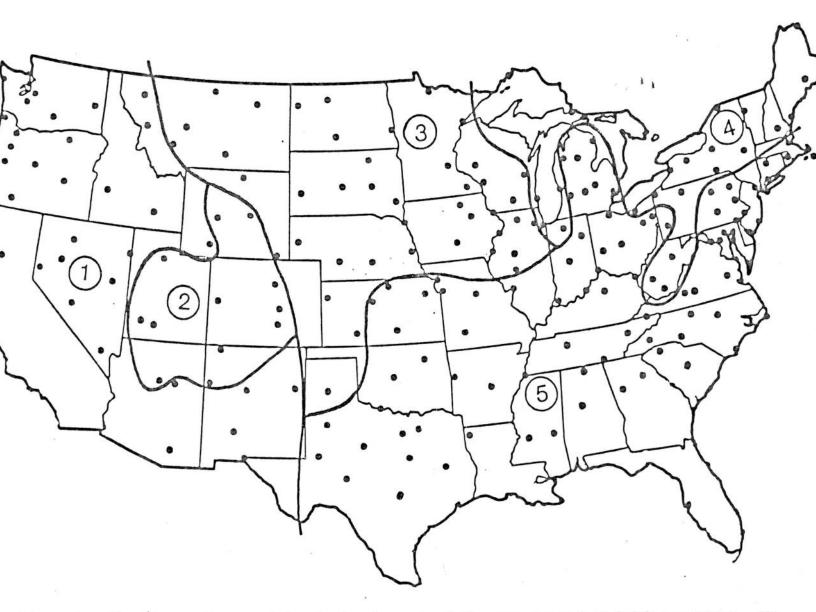


Figure 1. The five regions used in the development of the experimental PoSA(S) equations. The dots show the stations for which snow amount data were available in the developmental archive.